

Parameter Estimation of a dc Motor-Gear-ac Generator Mathematical Model

Abstract

Mathematical models and their parameters are essential for designers to predict the close loop behaviors of the plant so that systems are stable. A block model is developed in the MATLAB/simulink for the DC Motor-Gear-AC-Generator mathematical model in this paper, the block built is used to estimate the parameters in the estimation node using the gradient descent, simplex search and nonlinear least square algorithm. Gradient descent curve match that of the experimental data and its values are used in the DC Motor-Gear-AC Generator mathematical model. Key words: simulink, motor, generator, parameters.

Objective

- i) To built block simulink
- ii) Estimate the parameters of the DC Motor-Gear-Generator mathematical model.

1. Introduction

Mathematical models are useful for designers, engineers and mathematician to forecast system behaviors and system controllers. Knowledge on parameter estimation are important for describing a dc motor –gear-ac generator mathematical model in implementing an accurate mathematical model, designing precise controllers and predicting the closed loop behavior of the system. Parameter estimation can be done by the use of MATLAB/Simulink Data Acquisition Toolbox which allows the use of MATLAB as a single, integrated environment to support the entire data acquisition, data analysis, and application development process [1, 2]. Data Acquisition Toolbox supports Simulink with blocks model and allows verification and validation of data[3]. Parameters to be estimated are done as outlined by Koubaa [4]. The Simulink Parameter Estimation algorithm is used to set and estimate the system parameters[5], this algorithm support the creation of transient estimation, initial condition estimation, table values and simulation curves. The Simulink Parameter Estimation algorithm has inbuilt systems with ideal data which are related against the output data generated by the Simulink model[6]. By the use of optimization techniques, the software approximates the parameter and the initial conditions are stated in a way that the user-selected cost function is reduced[7]. The cost function characteristically calculates the least-square error between the model and the empirical data signals.

2. Motor- Gear-Generator Equation

The mathematical model for DC Motor-Gear-AC Generator is found using Kirchhoffs voltage law[8] Ohms law[9] and Newtons second law of motion[10]

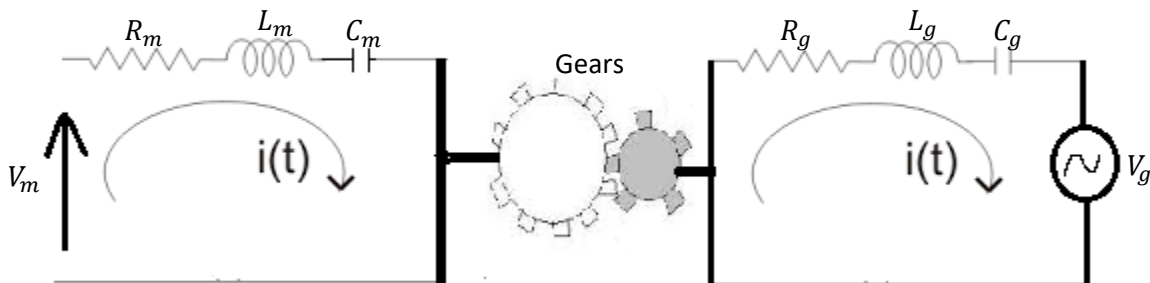


Figure 1: Schematic Representation of a DC Motor-AC Generator.[11]

The parameters of the equation to be estimated is the one discussed by Tarus, Koech and Obogi [11] in the equation below

$$\frac{V_g}{V_m} = Gr \left[\frac{C_m(C_gSK^2 - (R_gC_gS + L_gC_gS^2 + 1)(J_gS + B_g))}{C_g((R_mC_mS + C_mL_mS^2 + 1)(J_mS + B_m) + C_mSK^2)} \right] := G(s) \quad (1)$$

Or

$$G(s) = Gr \left[\frac{-AS^3 - BS^2 - CS - D}{HS^3 + GS^2 + FS + E} \right] \quad (2)$$

Where,

$$A = L_gC_gC_mJ_g, \quad B = R_gC_mC_gJ_g + L_gC_gC_mB_g, \quad C = C_mJ_g + C_mC_gR_gB_g - C_gC_mK^2$$

$$D = C_mB_g, \quad E = C_gB_m, \quad F = C_mC_gR_mB_m + C_mC_gK^2 + C_gJ_m$$

$$G = C_mC_gR_mJ_m + C_mC_gL_mB_m, \quad H = C_mC_gL_mJ_m$$

Equation (2) is the transfer function of DC Motor-AC Generator coupled Model

In control theory, a proper transfer function is a transfer function in which the degree of the numerator does not exceed the degree of the denominator [12]. This condition is satisfied by transfer function in equation (2) for the mathematical model derived by Tarus, Koech and Obogi.

3. Parameter estimation algorithm

Simulink estimation software have inbuilt techniques which include, the gradient descent which is a first order iterative optimization algorithm for finding the minimum of a function by finding a minimum error [13]. Nonlinear Least Square approach seeks to define the objective function that might reach its minimum [14]. Pattern Search Method can be utilized to determine parameters with insignificant error and compares it with the nonlinear least square method, pattern search operate by searching a set of point which expand or shrinks and the search stops after a minimum pattern size is reached [15]. Simplex Search Method involve a shift through a set of basic feasible solutions until the optimal basic feasible solution is identified whenever it exists. Before the model is run, numerical values are assign to each of the variables used in the model as the initial guess. For the dc Motor-ac Generator system the following values in Table 1

below are used. Figure 2 shows the Simulink block diagram for the system used alongside the throttle sensor inbuilt in the MATLAB and the parameter symbols with assign digit in the editor.

Table 1: Specification of the Motor-Generator Parameter Initial Values Used For Simulation.

No	Parameter Description (units)	Motor values	Generator values
1	Gear Ratio	1	1
2	Armature Resistance (Ω)	1	1
3	Inductance (H)	1	1
4	Viscous Friction Coefficients (Nm/(rad/s))	1	1
5	Moment of Inertial (kgm^2)	1	1
6	Capacitance (F)	1	1

shows the voltage response of the model using the initial parameter values in the model with a unit input voltage. Figure (4) show the required step response.

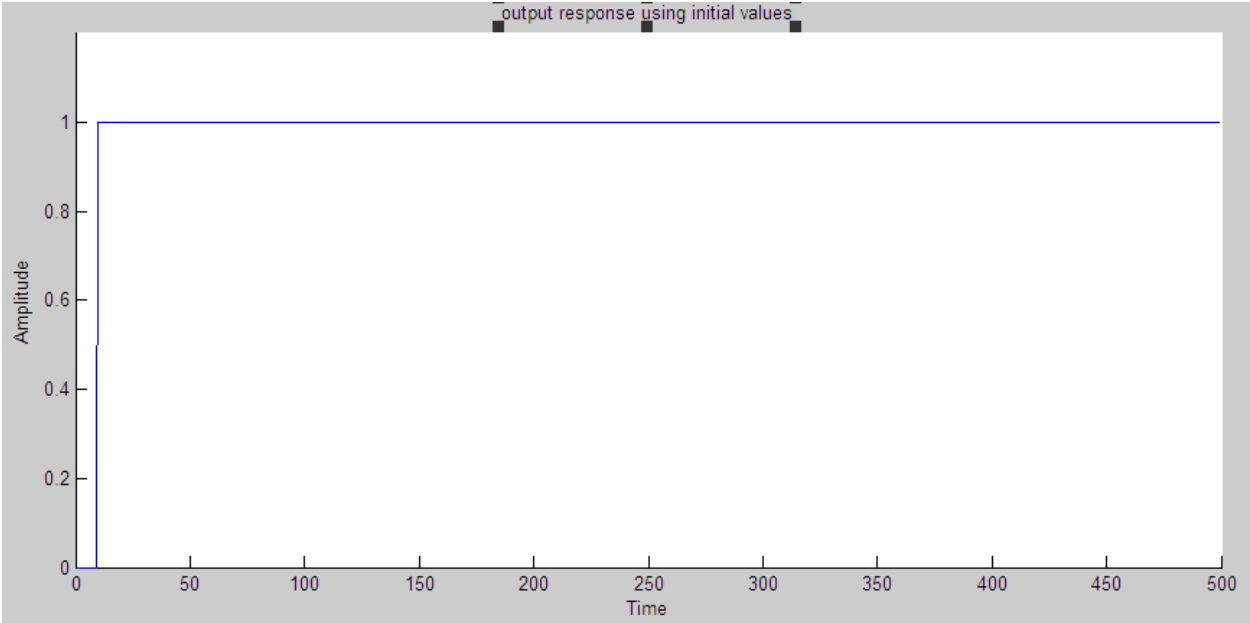


Figure 3: Output response of the Motor-Generator using initial values

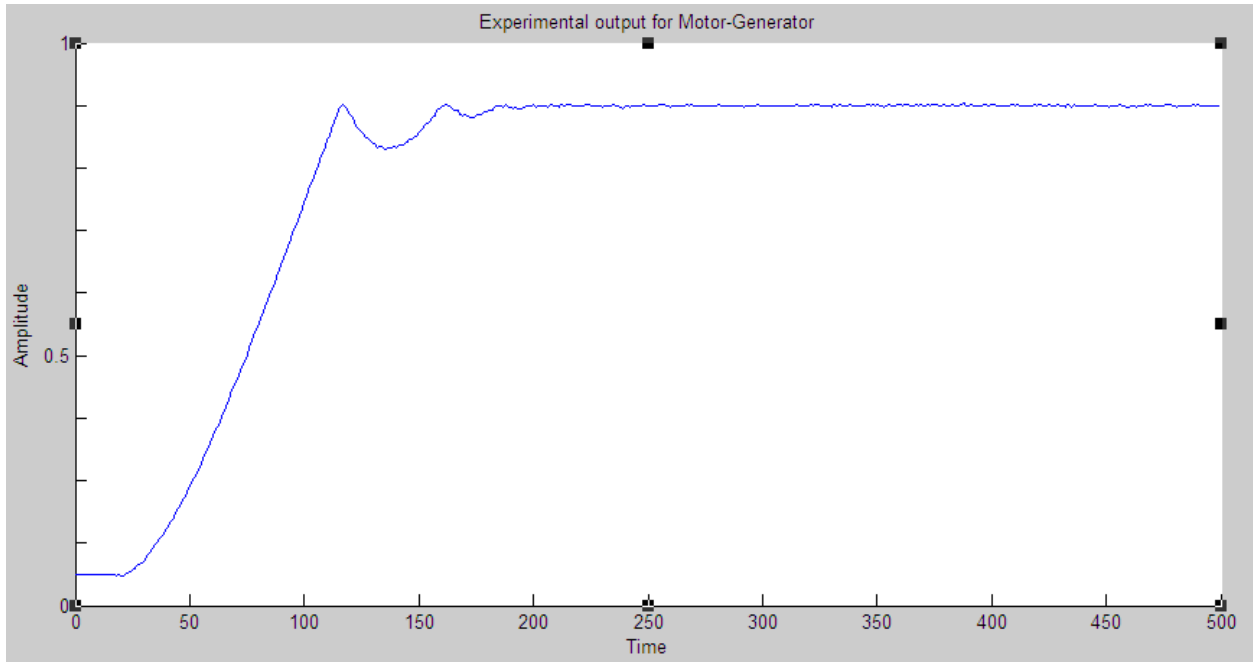


Figure 4: Expected Response of the Motor- Generator

4. Definition of Variables

Variables for estimation are stated to establish which parameters of the simulation can be adjusted and any rules governing their values. Eleven unknown parameters of the model are selected these parameters are the R_m Armature Resistance of the Motor, R_g Armature Resistance of the generator, J_m Motor Inertial, J_g Generator inertial, L_m Inductance of the motor, L_g Inductance of the Generator, C_m Capacitance of the Motor, C_g capacitance of the Generator B_m Viscous friction coefficients of the motor, B_g Viscous friction coefficients of the generator and the back e.m.f coefficient k . On the panel to the right of the list of parameters, the initial guesses for the parameter values are set and the minimum and maximum bounds on these values. The parameters of the system have positive values and the lower limits are set to zero shown in figure 5.

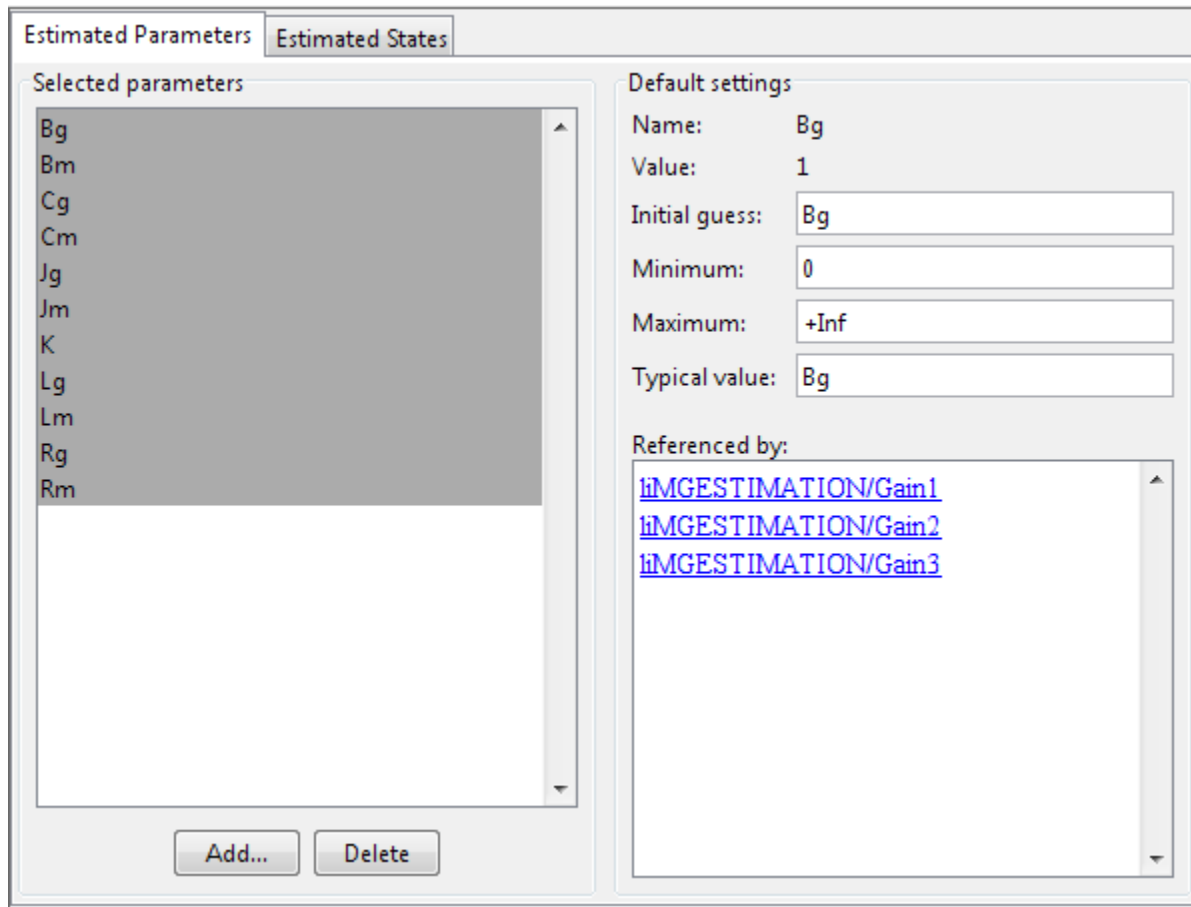


Figure 5 : Selecting Parameters for Estimation

In order to run estimation, an "Estimation" node is created. This is done on the "Estimation node and pressing the "New" button in the right-hand-side panel to create an estimation node called "New Estimation". The first panel is where data sets to be used in estimation is selected. It is possible to use one or more data sets at once in a given estimation. This model uses estimation data set. The next panel called "Parameters" is where parameters to be adjusted are selected for estimation. Eleven parameters are selected for estimation

Table 2: Parameters for Estimation

Data Sets Parameters States Estimation						
Estimation parameters						
Name	Value	Estimate	Initial Guess	Minimum	Maximum	Typical Va...
Bg	1	<input checked="" type="checkbox"/>	Bg	-Inf	+Inf	Bg
Bm	1	<input checked="" type="checkbox"/>	Bm	-Inf	+Inf	Bm
Cg	1	<input checked="" type="checkbox"/>	Cg	-Inf	+Inf	Cg
Cm	1	<input checked="" type="checkbox"/>	Cm	-Inf	+Inf	Cm
Jg	1	<input checked="" type="checkbox"/>	Jg	-Inf	+Inf	Jg
Jm	1	<input checked="" type="checkbox"/>	Jm	-Inf	+Inf	Jm
K	1	<input checked="" type="checkbox"/>	K	-Inf	+Inf	K
Lg	1	<input checked="" type="checkbox"/>	Lg	-Inf	+Inf	Lg
Lm	1	<input checked="" type="checkbox"/>	Lm	-Inf	+Inf	Lm
Rg	1	<input checked="" type="checkbox"/>	Rg	-Inf	+Inf	Rg
Rm	1	<input checked="" type="checkbox"/>	Rm	-Inf	+Inf	Rm

5. Results and Discussions

Two plot types are created to view the estimation results. The plot below shows the expected data against the simulated data. The simulated data come from the model with the estimated parameters. The results of the estimation appear satisfactory as the estimated (red) and (green) curve closely matches the measured results. Figure (6) shows voltage response after 100 iterations. The plot in Figure (7) shows the trajectory of the parameters at each iteration of the estimation process. It is shown that the parameters settle to their final values as the estimation process converges to a solution. Table (3), shows the results of Gradient descent algorithm, simplex search algorithm and nonlinear least square algorithm after 100 iteration

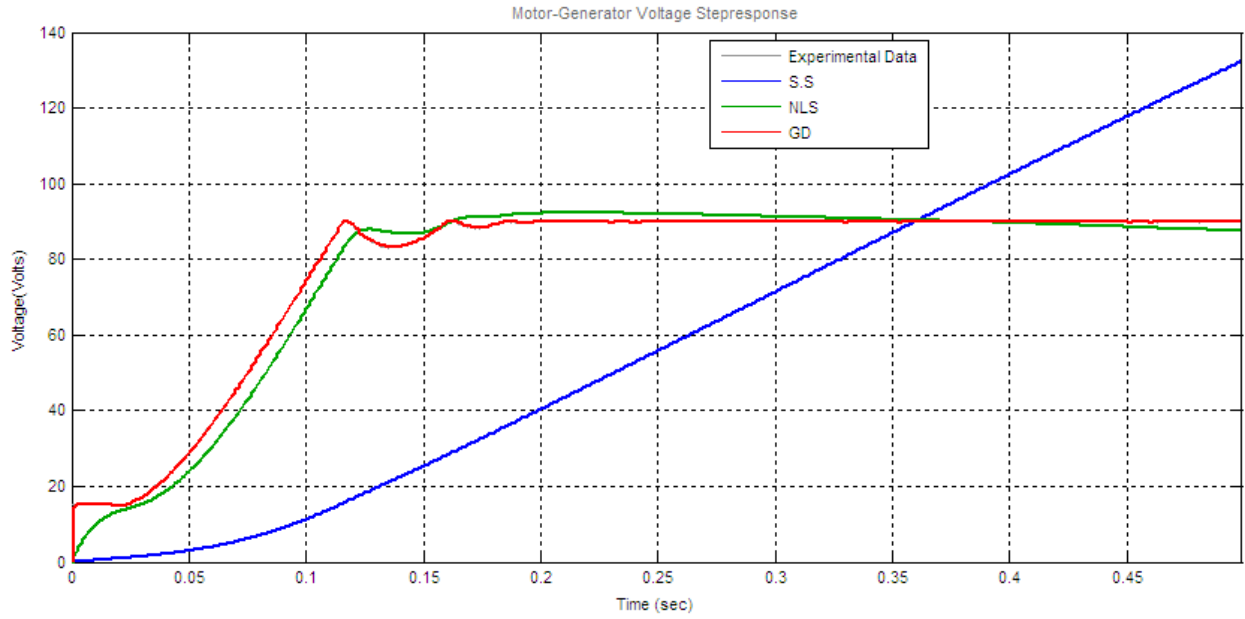


Figure 6 : Voltage Response for the Actual and Estimated Models

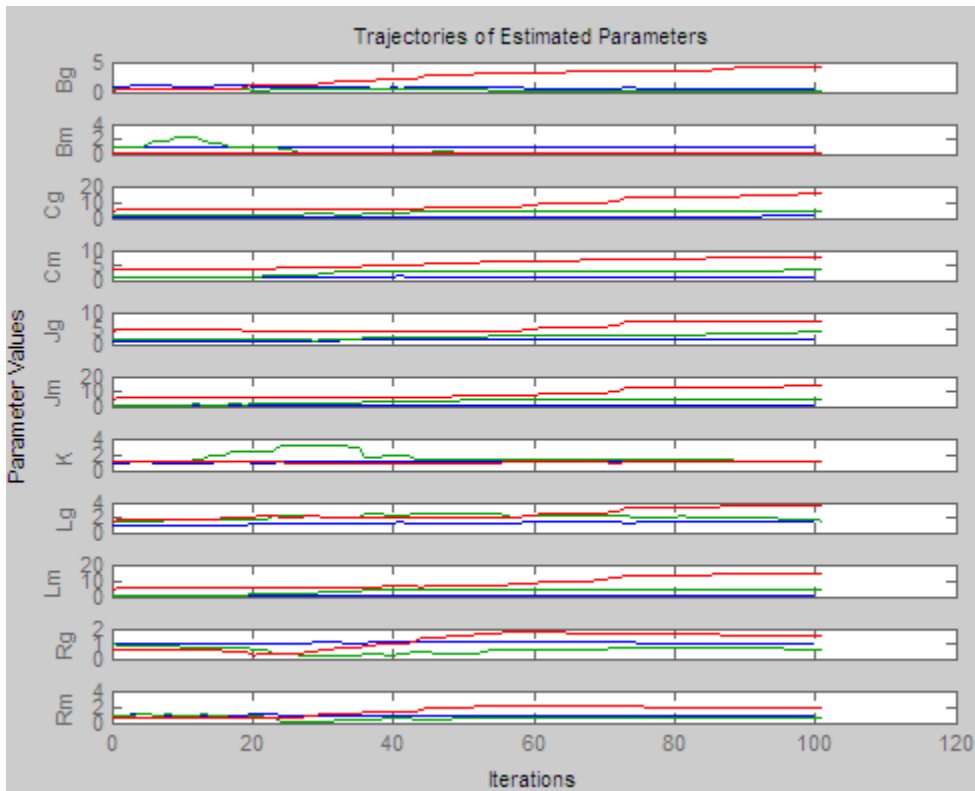


Figure 7 : Trajectories for Parameters

Table 3:Parameter Estimation Results

		Estimation on Three Algorithms					
Symbo	Name	Initial	Min	Max			
1		Guess	Set	Set	GD	NLS	S.S
			Value	Value			
B_g	Generator Viscous Friction	1	0	+infinity	4.137	0.012	0.545
B_m	Motor Viscous Friction	1	0	+infinity	0.013	0.004	0.844
C_g	Motor Capacitances	1	0	+infinity	15.055	4.649	1.288
C_m	Generator Capacitances	1	0	+infinity	7.867	3.384	1.304
J_g	Generator Moment of Inertia	1	0	+infinity	7.112	4.105	1.398
J_m	Motor Moment of Inertia	1	0	+infinity	13.898	4.935	0.966
K	Torque constant	1	0	+infinity	1.127	1.194	1.077
L_g	Generator Inductances	1	0	+infinity	3.635	1.576	1.371
L_m	Motor Inductances	1	0	+infinity	14.579	4.287	0.368

				y			
R_g	Generator Resistances	1	0	+infin	1.563	0.636	0.940
				y			
R_m	Motor Resistances	1	0	+infin	1.937	0.581	0.974
				y			

The results from the gradient descent are the best since the voltage response from the graph is more same to that of the experimental data and substituting the parameter values of the gradient descent into equation (2) the transfer function is

$$G(s) = Gr \left[\frac{-3062.05023s^3 - 3097.853178s^2 - 671.426095s - 32.54648418}{23996.498s^3 + 3210.1686s^2 + 362.5291438s + 0.188413325} \right] \quad (3)$$

Equation (3) is the transfer function of DC Motor- Gear-AC Generator model.

6. Estimated Parameters of the DC Motor-AC Generator System

The tables (4) and (5) shows specific parameters of the DC Motor-AC Generator system as obtain from the MATLAB/Simulink using gradient descent algorithm.

Table 4 : Motor specification

Gr	$C(F)$	$V(V)$	$B \frac{Nm}{rad}$ /sec	$J(Kgm^2)$	K	$L(H)$	$i(A)$	$R(\Omega)$	$T(Nm)$	$\omega(rad/s)$
1	7.87	12	0.01	13.90	1.13	14.58	0.09	1.94	0.11	10.62
2	7.87	12	0.01	13.90	1.13	14.58	0.09	1.94	0.11	10.62
3	7.87	12	0.01	13.90	1.13	14.58	0.09	1.94	0.11	10.62
4	7.87	12	0.01	13.90	1.13	14.58	0.09	1.94	0.11	10.62
5	7.87	12	0.01	13.90	1.13	14.58	0.09	1.94	0.11	10.62
6	7.87	12	0.01	13.90	1.13	14.58	0.09	1.94	0.11	10.62
7	7.87	12	0.01	13.90	1.13	14.58	0.09	1.94	0.11	10.62

Table 5 : Generator specification

<i>Gr</i>	$\frac{Nm}{B} (\frac{rad}{s})$	<i>C(F)</i>	<i>J(Kgm²)</i>	<i>K</i>	<i>L(H)</i>	<i>R(Ω)</i>	$\omega(rad/s)$	<i>V(V)</i>	Excess Voltage
1	4.14	15.06	7.11	1.13	3.64	15.06	10.62	12.00	0
2	4.14	15.06	7.11	1.13	3.64	15.06	21.24	24.00	12
3	4.14	15.06	7.11	1.13	3.64	15.06	31.86	36.00	24
4	4.14	15.06	7.11	1.13	3.64	15.06	42.48	48.00	36
5	4.14	15.06	7.11	1.13	3.64	15.06	53.10	60.00	48
6	4.14	15.06	7.11	1.13	3.64	15.06	63.72	72.00	60
7	4.14	15.06	7.11	1.13	3.64	15.06	74.41	84.00	72

7. Conclusion

Three method of parameter estimation were created in the transient data node of the MATLAB Simulink, these include gradient descent, nonlinear least square and the simplex search methods. The curve of the gradient descent converges to that of experimental data curve making it the appropriate for the system. The estimated parameters are used to generate simulation curves for voltage and time step responses. These curves agree with agreement with actual curves within the precision level of the model. The values obtained from the gradient descent are substituted to the unknown values of the formulated equation. Identification process was built and done using the MATLAB. More work can be done using microcontrollers controllers and digital signal controllers.

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